# Data based correction of MC

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#### Introduction

- Investigate two methods of correcting simulation based on data:
- With data interpolation:
  - Interpolate data in fine  $(p, \theta)$  bins, preserving integrals in the original data bins from the publication
  - 2 Divide interpolated data by MC in fine bins  $\rightarrow$  correction factor
  - 1 Draw spectrum in  $(x_F, p_T)$  bins, weighting each track by the correction factor
- Without data interpolation:
  - lacktriangle Divide data by MC in original bins  $\rightarrow$  correction factor
  - ② Draw spectrum in  $(x_F, p_T)$  bins, weighting each track by the correction factor
- Test data: \(\pi^-\) spectra produced in p+p interactions at 80 GeV/c generated with EPOS and VENUS models, 5M interactions each.

In this presentation I pretend EPOS is "data" and VENUS is "MC"

# Procedure without data interpolation

• Correction factor is obtained by dividing data by MC in the same  $(p, \theta)$  bins as in which the data was published:

$$c(p, \theta) = \frac{\mathtt{data}(p, \theta)}{\mathtt{MC}(p, \theta)}$$
 (1)

 Data spectrum in (x<sub>F</sub>, p<sub>T</sub>) can be obtained by filling histogram with the same MC tracks using corrections obtained in the previous step:

$$data(x_{\mathsf{F}}, p_{\mathsf{T}}) = \mathtt{MC}(x_{\mathsf{F}}, p_{\mathsf{T}}) \cdot c(p, \theta) = \mathtt{MC}(x_{\mathsf{F}}, p_{\mathsf{T}}) \cdot \frac{data(p, \theta)}{\mathtt{MC}(p, \theta)}. \tag{2}$$

• The procedure takes shape of the MC distribution and scales it by a data-based correction factor in each bin.

# Procedure with data interpolation

- Interpolating data data $(p, \theta)$  results in data in fine binning data $^f(p, \theta)$
- Correction factor is obtained by dividing it by MC in fine binning:

$$c(p,\theta) = \frac{\text{data}^f(p,\theta)}{\text{MC}^f(p,\theta)}$$
(3)

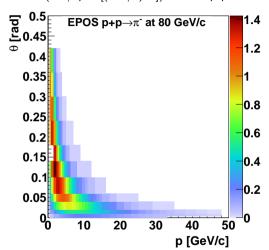
• Data spectrum in  $(x_F, p_T)$  can be obtained by filling histogram with the same MC tracks using corrections obtained in the previous step:

$$\operatorname{data}(x_{\mathsf{F}}, p_{\mathsf{T}}) = \operatorname{MC}(x_{\mathsf{F}}, p_{\mathsf{T}}) \cdot c(p, \theta) = \operatorname{MC}(x_{\mathsf{F}}, p_{\mathsf{T}}) \cdot \frac{\operatorname{data}^{f}(p, \theta)}{\operatorname{MC}^{f}(p, \theta)} . \tag{4}$$

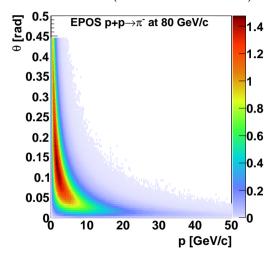
Q: Why do we need MC at all? Note that even very fine bins in  $(p, \theta)$  space can grow large in  $(x_F, p_T)$  space. If we didn't use MC it would be equivalent to assumption that the data is perfectly flat within the fine bin

## EPOS "data"

data  $(d^2n/dpd\theta [(GeV/c)^{-1}])$  in NA61 paper bins



data in fine bins (in real world we can't see it)

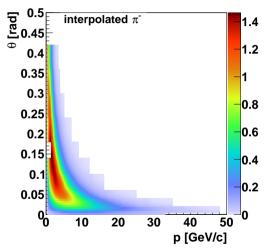


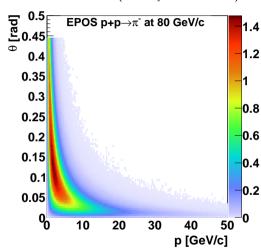
# EPOS "data" interpolated





data in fine bins (normally we can't see it)



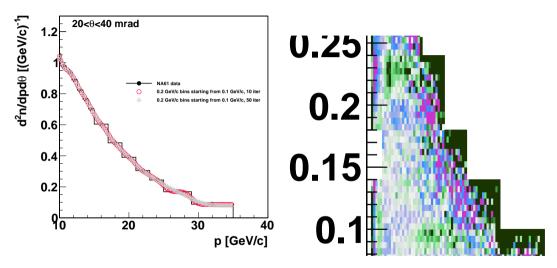


## Interpolation error

(interpolated data) / (data in fine bins) same plot, color scale zoomed interpolated "data" 1.5 De 0.5 0.45 interpolated "data" 1.1 0.5 0.45 hi-res "data" hi-res "data" 1.4 1.08 0.4 1.3 0.4 1.06 0.35 1.2 0.35 1.04 0.3 1.1 0.3 **−1.02** 0.25 0.25 0.2 0.9 0.2 0.98 0.15 0.8 0.15 0.96 0.1 0.7 0.1 0.94 0.05 0.05 0.92 0.6 0.5 O<sub>r</sub> 0.9 50 50 10 20 30 10 20 30 40 40 p [GeV/c] p [GeV/c]

- Large errors at the edges, 10% biases in the whole range
- Of course the errors would be much larger with no interpolation. Also the integral is preserved

# What happens at the edges?



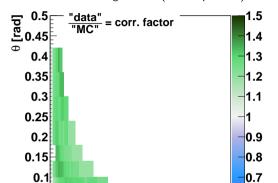
- There is no data to interpolate at the edges, so we begin to extrapolate. Extrapolation requires assumptions.
- ullet Notice how much less error is in the "dent" area at heta=0.15 rad

## Correction factor

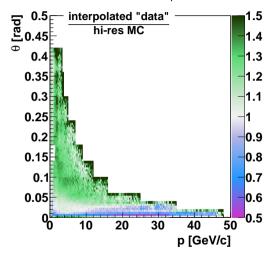
0.05

0

correction in original bins (no interpolation)



correction based on interpolated data



10

20

30

40

p [GeV/c]

0.6

0.5

50

MC in  $(x_F, p_T)$ 

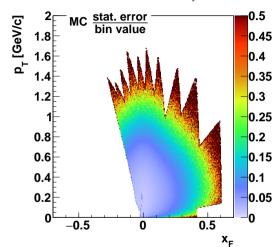
0.2

0



## 30 $p_{_T}[GeV/c]$ MC 1.8 25 1.6 1.4 20 1.2 15 8.0 10 0.6 0.4 5

#### relative statistical uncertainty of MC



Plotted only in the region covered by the "data" bins

0.5

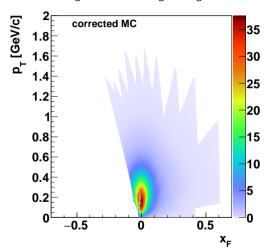
 $\mathbf{X}_{\mathsf{F}}$ 

• Statistical errors significant at the edges

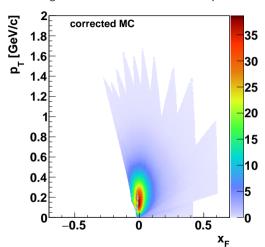
-0.5

# Corrected MC in $(x_F, p_T)$

using correction in original large bins



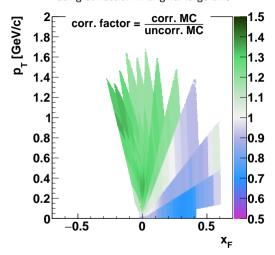
using correction in fine bins from interpolation



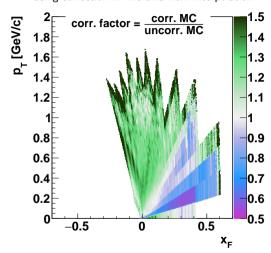
• For each particle a correction weight was applied based on its  $(p, \theta)$  coordinates

#### Effective correction factor

using correction in original large bins

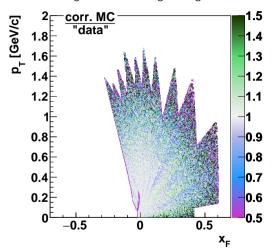


using correction in fine bins from interpolation

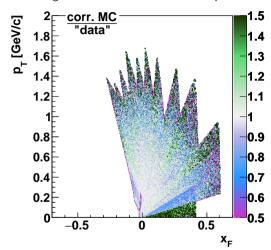


#### Bias of corrected data

using correction in original large bins



using correction in fine bins from interpolation



• Statistical fluctuations need to be disregarded in these plots

# Summary

#### Comparison of two methods

- ullet Both method introduce some 10% systematic biases here and there o expected as we have data in large bins only
- Interpolation introduces large errors close to some edges

#### Possible improvements

- Interpolation method
  - lacktriangle Improve the interpolation method ightarrow tried already, little improvements
  - ightharpoonup Omit bins at the edges from the analysis ightharpoonup waste of data
  - ▶ Manually add fake data bins at the edges to improve interpolation → lot's of work, introduces model dependence, difficult to defend
- No interpolation method
  - Interpolate the correction factor → 1. I'm not yet sure how 2. risk of running into the same issues with interpolation 3. But possibly interpolating the correction factor introduces less error than interpolating the spectrum?
- Ask model creators to tune their models
  - ▶ They may have much better experience in solving these kind of problems than us

### Other future steps

• Each method requires more testing with various test data sets to estimate size of bias it introduces